

A multivariate statistic approach to detect process imbalance in lab scale continuously stirred tank anaerobic reactors submitted to inadequate organic loading rate

Sébastien Lemaigre¹, G. Adam², X. Goux¹, P. Delfosse¹

¹Département EVA, Centre de Recherche Public Gabriel Lippmann, 41 rue du Brill, L-4422 Belvaux, Luxembourg
² Université de Liège, Arlon Campus Environnement, 185 Avenue de Longwy, B-6700 Arlon, Belgium

INTRODUCTION

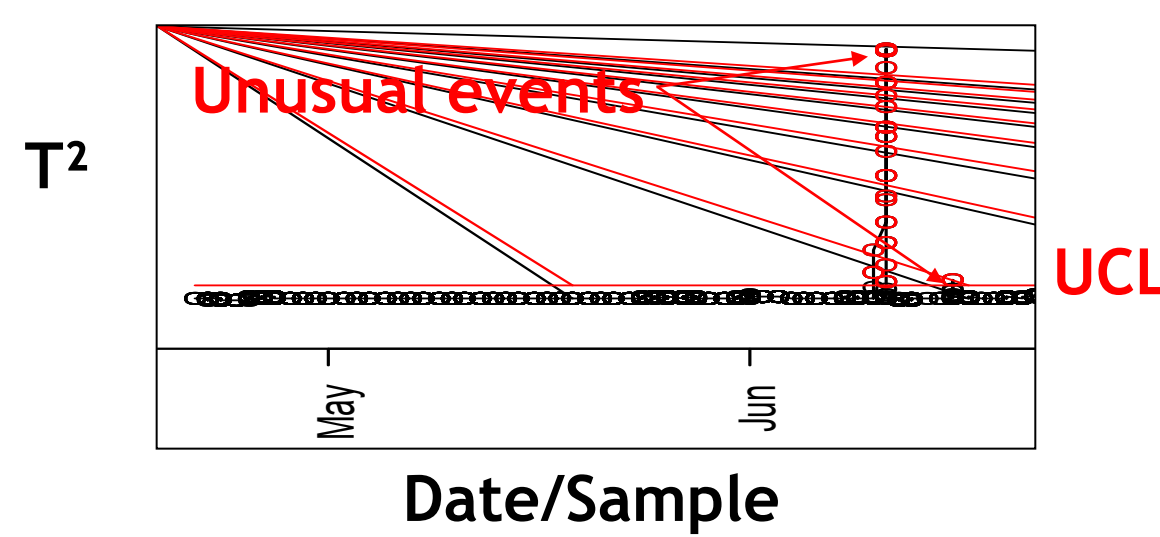
During recent decades, anaerobic digestion (AD) of organic substrates has become one of the most mature technologies to produce renewable energy from wet biomass. Biogas production using AD process implies to manage a complex microbial flora that is highly sensitive to variations of the ecological conditions existing in the reactors. The main biological dysfunctions are nowadays better understood and are for the most part related to inappropriate feeding of the digesters. So far various methods have been evaluated to monitor the process but none seems to be ideal. These methods usually consist in measuring a set of variables judged to be characteristic of the process status (i.e. pH of the liquid phase, CH₄/CO₂ ratio of the biogas,...) and interpreting the collected data for each parameter individually. However, since these variables reflect the conditions of the reactor anaerobic microbial community, it appears very probable that they present a certain degree of correlation. An efficient tool for AD process monitoring should therefore benefit from the integration of information about the way the measured parameters interact when the process is in control. A way to satisfy this condition is to monitor the reactors using multivariate statistic methods as an alternative to usual univariate approaches. Control charts built with Hotelling T² statistic are well adapted to this kind of application. Those charts are becoming more and more popular to monitor continuous processes where the quality of the final product is related to several dependant variables (Adam et al., 2012).

AIMS of the STUDY

Assessing the potential of Hotelling's multivariate control charts built from variables reputed as individual AD process indicators to detect dysfunction of anaerobic reactors submitted to inadequate feeding conditions.

Hotelling's control charts?

- The most applied chart in multivariate process control.
- Phase I:** Data set built with measurements performed during a period where the process is supposed to be "in control". It authorizes to compute 2 parameters that characterize this in-control situation: the covariance matrix (S) and the mean vector (X_{mv}).
- Phase II:** Each collected data sample is used to compute a unique statistical parameter (T²) that compare the current sample data structure with Phase I X_{mv} and S. T² value are compared to an upper control limit (UCL) to identify out-of-control situations:



MATERIAL AND METHODS

Overfeeding campaigns with lab scale continuous reactors

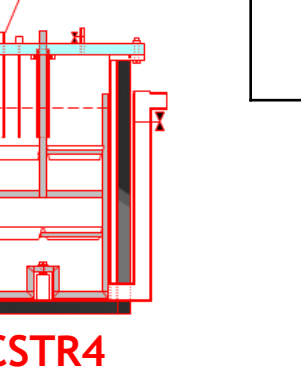
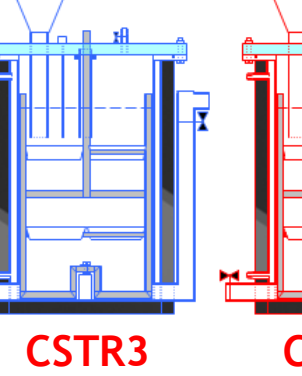
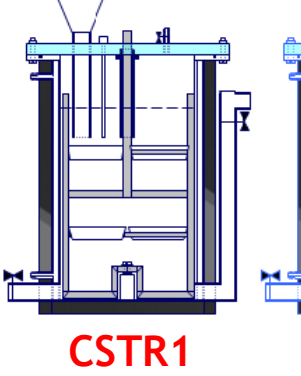
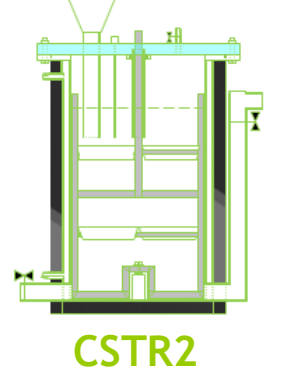
Continuous data collection

Analysis of the data sets with Hotelling's control charts

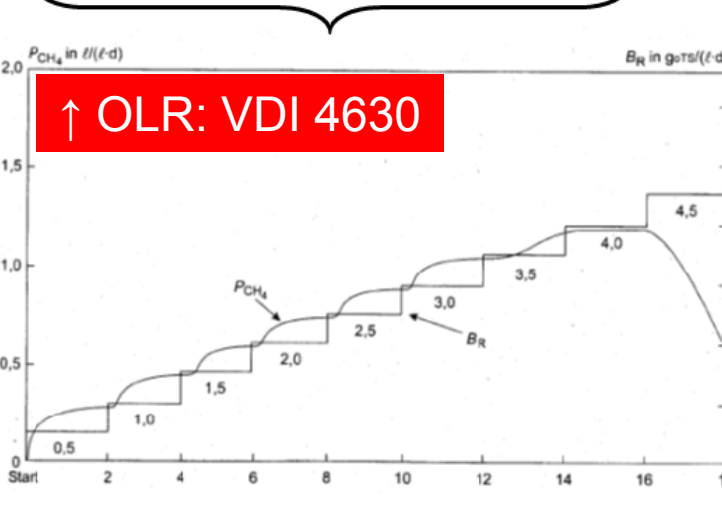
4 Continuously Stirred Tank Reactors (CSTR) 100l
Mesophilic temperature range (37°C)

1 control reactor - cautious organic loading rate (OLR)

3 overfed reactors



Inoculum: sludge from the anaerobic digester of a waste water treatment plant (mesophilic conditions).
Substrate: dried sugar beet pulps.

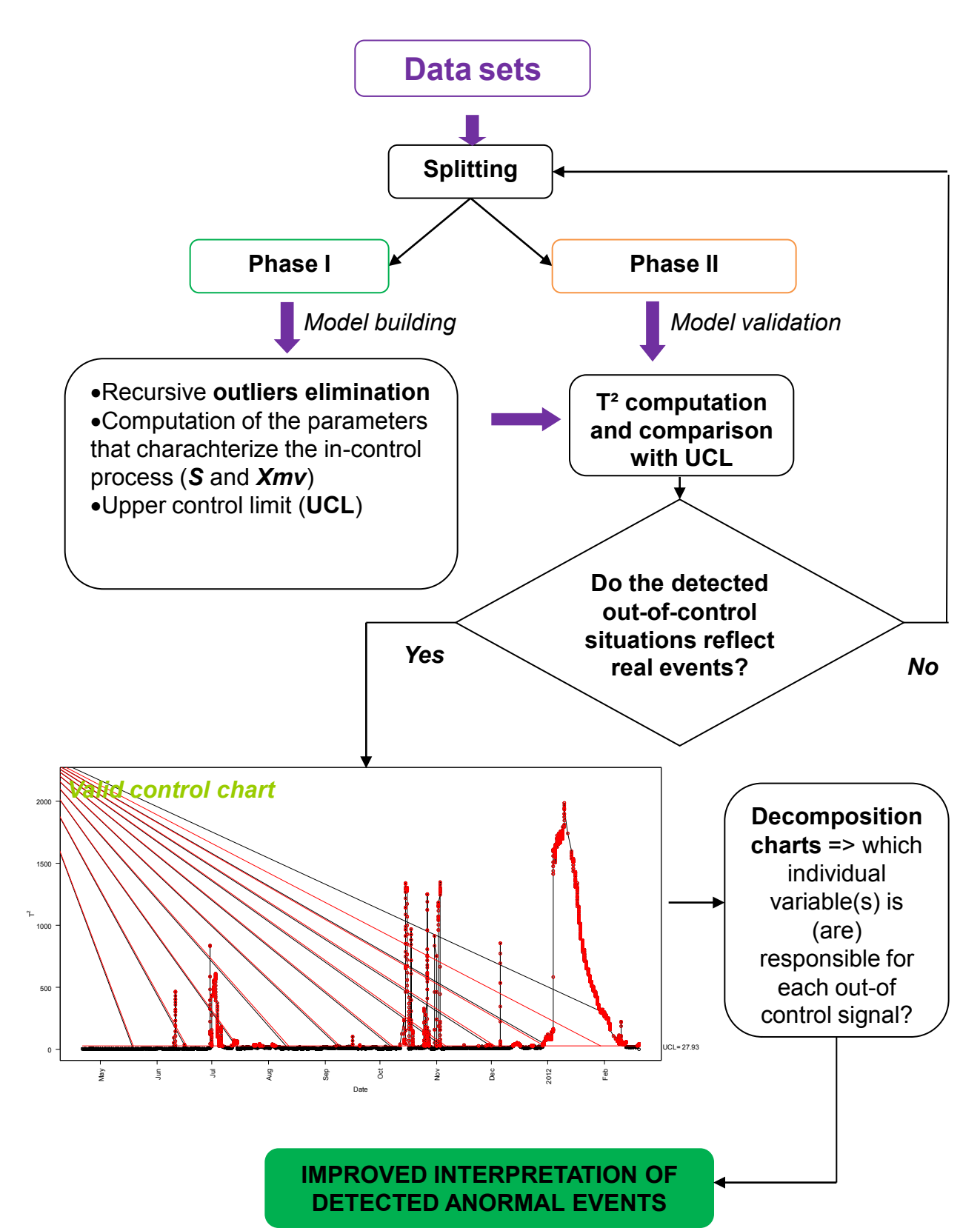


In the biogas :

Parameter	Units	Measurement method	Measurement Frequency
Biogas specific production	l/h	drum-type wet gas meter	h ⁻¹
CH ₄ concentration	% (Volume)	Non Dispersive Infra-Red Sensor (NDIR)	(2h) ⁻¹
CO ₂ concentration	% (Volume)	NDIR	(2h) ⁻¹
H ₂ concentration	ppm (Volume)	Metal Oxide Semi-conductor (MOS) with molecular sieve	(2h) ⁻¹
H ₂ S concentration	ppm (Volume)	Electro-Chemical Sensor (ECS)	(2h) ⁻¹

In the digestate :

Parameter	Measurement method	Measurement Frequency
pH	Saturated calomel electrode	d ⁻¹
Total solids (TS)	Gravimetry (VDI 4630)	w ⁻¹
Volatile solids (VS)	Gravimetry (VDI 4630)	w ⁻¹
Total alkalinity (TA)	Volumetric (BiogasPro, Germany)	w ⁻¹
Ammoniac Nitrogen (NH ₄ -N)	Volumetric (BiogasPro, Germany)	w ⁻¹



*Control charts were built using « MSQC » R package v1.0.1 (E. Santos-Fernandez)
 *Decomposition chart were computed using « MYT » method (Mason et al., 1995)

RESULTS and DISCUSSION

1) Selection and optimization of the phase I data set (model building)

Phase I (T² after outliers cleaning)

Phase II (CSTR1)

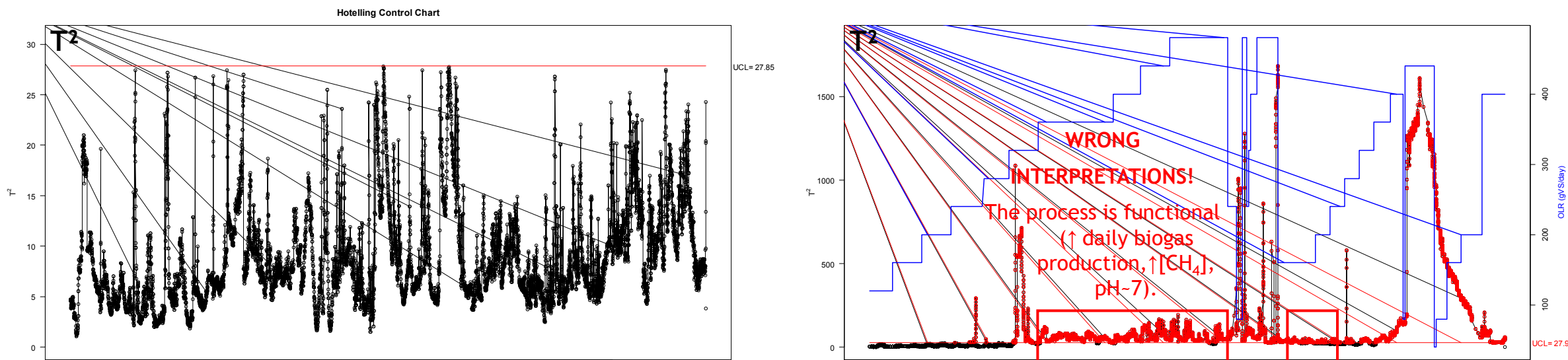


Fig. 1. All the measured parameters are used for model building except pH. Phase I is performed with a dataset issued from the cautiously fed control reactor (CSTR2). In phase II, an excessive number of samples are interpreted as out-of-control situations even though the digestion process is not perturbed.

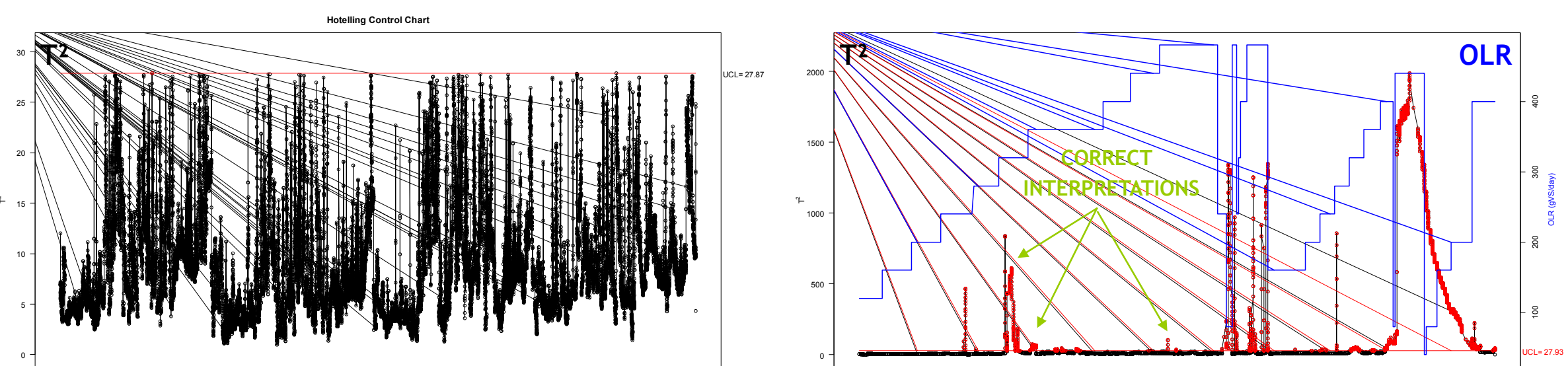


Fig. 2. All the measured parameters are used for model building except pH. Phase I is performed with a dataset combining mixed data from the 4 reactors. In phase II, the detected unusual events can be related to real process dysfunction or real failures of the measurement system (cf. 2).

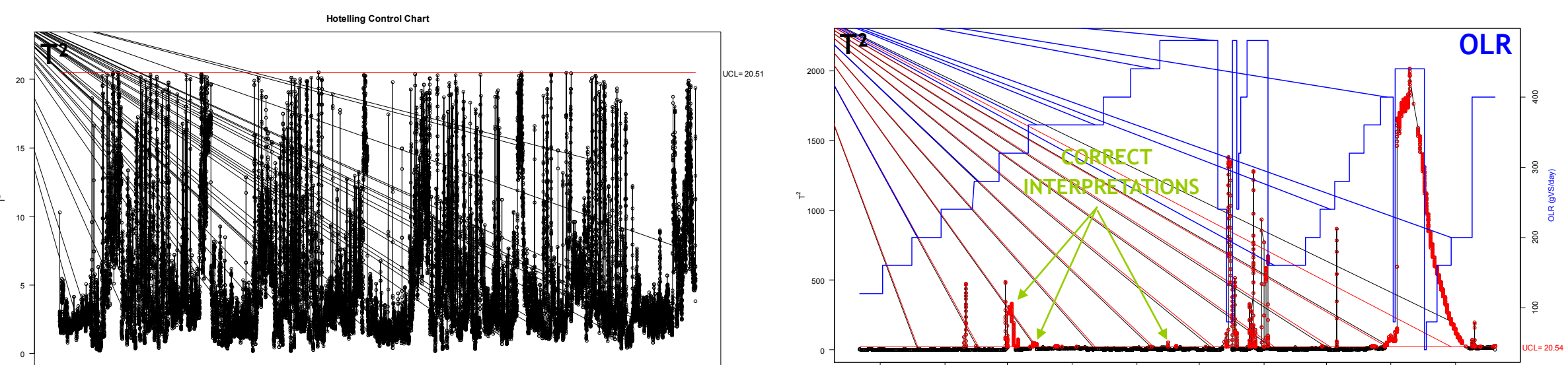
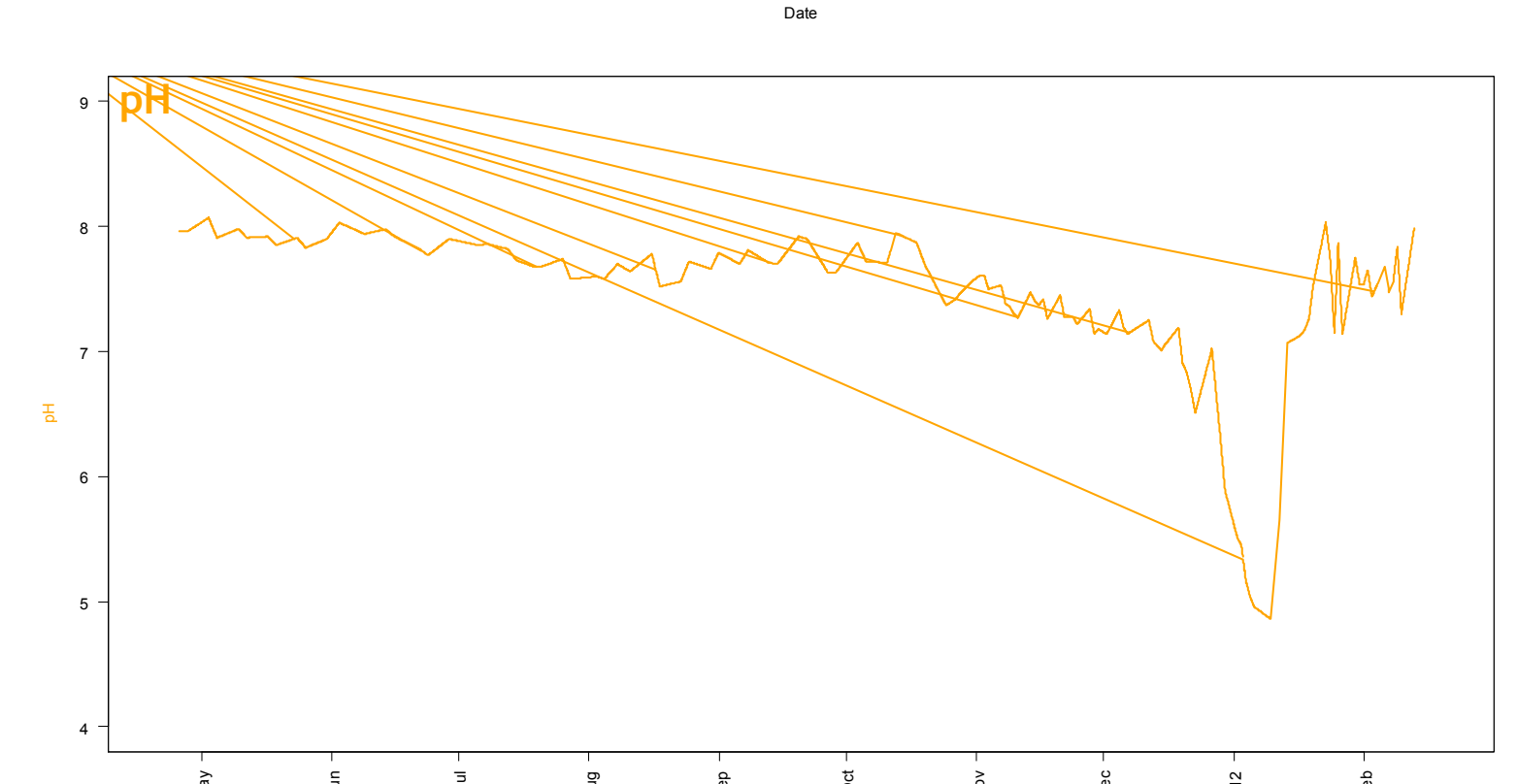
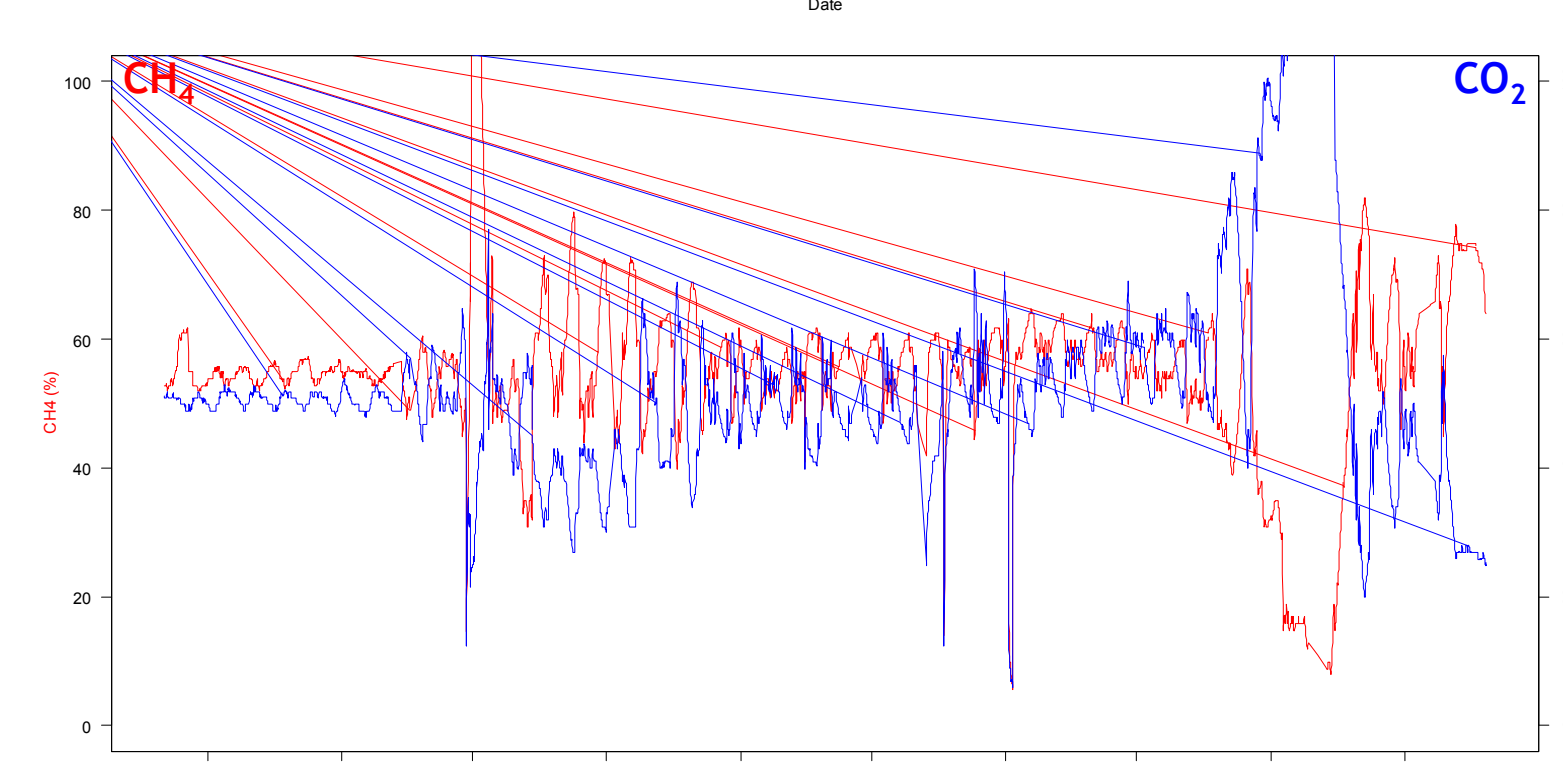
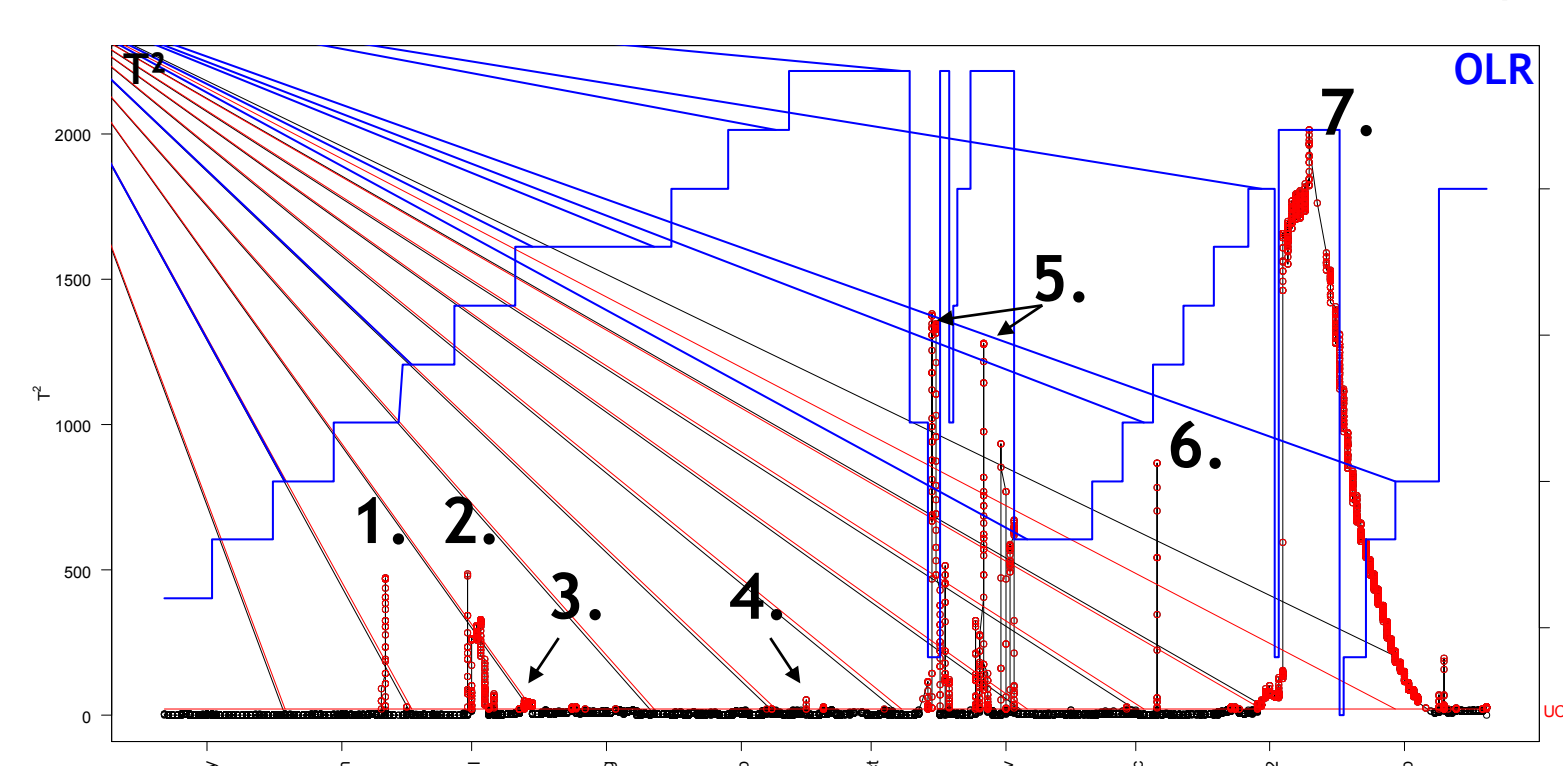


Fig. 3. Phase I is performed with a dataset combining mixed data from the 4 reactors but only the parameters measured in the gas phase are used. Phase II still provide correct interpretation of the unusual events.

2) Process control for 1 overfed reactor (CSTR1)

NB : data set used for phase I was described Fig. 3. pH of liquid phase is NOT included in the data sets.



- Event 1 and 6: A stirrer malfunction occurred causing an important hydrogen production. Decomposition showed that hydrogen concentration in the biogas was the parameter responsible for the detected out-of-control situation (Fig. 4).
- Event 2: An important foam generation caused a failure of the gas sensors and required the disconnection of the biogas analyser for maintenance.
- Event 3: Drift of CH₄ and CO₂ sensors.
- Event 4: H₂S sensor was used to analyse a gas sample extremely concentrated. The sensor stayed perturbed during some hours after the measurement.
- Event 5: Unexplained hydrogen production peaks.
- Event 7: The reactor reaches an acidosis situation. The first out-of-control samples are related to an unusual relationship between CH₄ and CO₂ concentrations. Highest T² values are related to increasing hydrogen concentration.

Combinations of variables	T ² decomp	ucl	p-value
CH ₄	0.17	10.83	0.878
CO ₂	0.06	10.83	0.813
H ₂	374.20	> 10.83	0
Biogas sp. pr.	0.27	10.83	0.804
H ₂ S	0.21	10.83	0.845
CH ₄ CO ₂	0.18	13.83	0.837
CH ₄ H ₂	374.49	> 13.83	0
CH ₄ Biogas sp. pr.	0.32	13.83	0.728
CH ₄ H ₂ S	0.42	13.83	0.656
CO ₂ H ₂	378.77	> 13.83	0
CO ₂ Biogas sp. pr.	0.28	13.83	0.757
CO ₂ H ₂ S	0.22	13.83	0.799
H ₂ Biogas sp. pr.	410.38	> 13.83	0
H ₂ H ₂ S	384.72	> 13.83	0
Biogas sp. pr. H ₂ S	0.59	13.83	0.566
CH ₄ CO ₂ H ₂	385.62	> 16.28	0
CH ₄ CO ₂ Biogas sp. pr.	0.32	16.28	0.810
CH ₄ CO ₂ H ₂ S	0.55	16.28	0.650
CH ₄ H ₂ Biogas sp. pr.	422.83	> 16.28	0
CH ₄ H ₂ H ₂ S	385.40	> 16.28	0
CH ₄ Biogas sp. pr. H ₂ S	0.54	16.28	0.590
CO ₂ H ₂ Biogas sp. pr.	410.78	> 16.28	0
CO ₂ H ₂ H ₂ S	395.40	> 16.28	0
CO ₂ Biogas sp. pr. H ₂ S	0.60	16.28	0.616
H ₂ Biogas sp. pr. H ₂ S	416.16	> 16.28	0
CH ₄ CO ₂ H ₂ Biogas sp. pr.	439.77	> 18.49	0
CH ₄ CO ₂ H ₂ H ₂ S	422.20	> 18.49	0
CH ₄ CO ₂ Biogas sp. pr. H ₂ S	0.79	18.49	0.533
CH ₄ H ₂ Biogas sp. pr. H ₂ S	429.11	> 18.49	0
CO ₂ H ₂ Biogas sp. pr. H ₂ S	418.73	> 18.49	0
CH ₄ CO ₂ H ₂ Biogas sp. pr. H ₂ S	486.64	> 20.51	0

Fig. 4. Decomposition chart for event 1. All combinations of variables including hydrogen are responsible for the alarm.

CONCLUSIONS

- Hotelling's T² control charts appear as an interesting tool for AD process monitoring. The study showed that it was possible to quickly detect out-of-control situations (related to both biological dysfunctions and measurement system failure) using only gas phase parameters to build the model.
- T² is a unique parameter => the charts are easy to read and the T² value can be easily used to trigger various events to inform the manager of the reactors (i.e. alarm signal).
- Using the data of a reactor that was cautiously and steadily fed to define the in-control situation of the process did NOT appear as the best option. It was shown as more efficient to add to the training data set the measurements from overfed reactors. After cleaning of the outliers, the data set included more information on how the variables interact each others for a wider range of OLR values. Subsequent event detection was then more realistic.
- In this study, liquid phase measurements did not seem to provide important added-value to the process control tool, probably because of their low frequency of measurement compared to biogas parameters.

Adam, G., Lemaigre, S., Romain, A. C., Nicolas, J., & Delfosse, P. (2013). Evaluation of an electronic nose for the early detection of organic overload of anaerobic digesters. *Bioprocess and biosystems engineering*, 36(1), 23-33
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Contact: Sébastien Lemaigre, CRP - Gabriel Lippmann 41, rue du Brill - L-4422 BELVAUX Tél. (+352) 47 02 61-458 lemaigre@lippmann.lu